

# Impact of Severe Weather Events on Public Health and Economy in the United States.

## Synopsis

The U.S. National Oceanic and Atmospheric Administration's (NOAA) storm database tracks characteristics of major storms and weather events in the United States, include when and where they occur, as well as estimates of any fatalities, injuries and property damage. This report contains the exploratory analysis results on the health and economic impact by the severe weather events based on the data from NOAA database.

## Required packages

```
library(R.utils)
library(ggplot2)
library(gridExtra)
library(knitr)
```

## Loading data in R

Download and unzip the data in the working directory.

```
if (!"stormData.csv.bz2" %in% dir(".")) {
  download.file("https://d396qusza40orc.cloudfront.net
    /repdata%2Fdata%2FStormData.csv.bz2"
    , destfile = "stormData.csv.bz2")
  bunzip2("stormData.csv.bz2", overwrite=TRUE, remove=FALSE)
}
```

Load the data.

```
data<-read.csv("stormData.csv", sep=",")
head(data)
```

```
##   STATE__      BGN_DATE BGN_TIME TIME_ZONE COUNTY COUNTYNAME STATE
## 1      1 4/18/1950 0:00:00    0130      CST    97    MOBILE    AL
## 2      1 4/18/1950 0:00:00    0145      CST     3    BALDWIN   AL
## 3      1 2/20/1951 0:00:00    1600      CST    57    FAYETTE    AL
## 4      1 6/8/1951 0:00:00    0900      CST    89    MADISON    AL
## 5      1 11/15/1951 0:00:00    1500      CST    43    CULLMAN    AL
## 6      1 11/15/1951 0:00:00    2000      CST    77 LAUDERDALE  AL
##   EVTYPE BGN_RANGE BGN_AZI BGN_LOCATI END_DATE END_TIME COUNTY_END
## 1 TORNADO         0         0          0      0      0      0
## 2 TORNADO         0         0          0      0      0      0
## 3 TORNADO         0         0          0      0      0      0
## 4 TORNADO         0         0          0      0      0      0
## 5 TORNADO         0         0          0      0      0      0
## 6 TORNADO         0         0          0      0      0      0
##   COUNTYENDN END_RANGE END_AZI END_LOCATI LENGTH WIDTH F MAG FATALITIES
```

```
## 1      NA      0      14.0  100 3  0      0
## 2      NA      0      2.0  150 2  0      0
## 3      NA      0      0.1  123 2  0      0
## 4      NA      0      0.0  100 2  0      0
## 5      NA      0      0.0  150 2  0      0
## 6      NA      0      1.5  177 2  0      0
##  INJURIES  PROPDMG  PROPDMGEXP  CROPDGMG  CROPDGMGEXP  WFO  STATEOFFIC  ZONENAMES
## 1      15     25.0      K      0
## 2       0      2.5      K      0
## 3       2     25.0      K      0
## 4       2      2.5      K      0
## 5       2      2.5      K      0
## 6       6      2.5      K      0
##  LATITUDE  LONGITUDE  LATITUDE_E  LONGITUDE_  REMARKS  REFNUM
## 1      3040      8812      3051      8806      1
## 2      3042      8755       0      0      2
## 3      3340      8742       0      0      3
## 4      3458      8626       0      0      4
## 5      3412      8642       0      0      5
## 6      3450      8748       0      0      6
```

## Data Processing

Calculate actual "PROPERTY\_DAMAGE" and "CROP\_DAMAGE" in billions US dollars and add them as columns in the dataset.

```
levels(data$PROPDMGEXP)
```

```
## [1] ""  "-"  "?"  "+"  "0"  "1"  "2"  "3"  "4"  "5"  "6"  "7"  "8"  "B"  "h"  "H"  "K"
## [18] "m"  "M"
```

```
levels(data$PROPDMGEXP)<-c(1,0,0,0,1,10,100,1000,1e+04,1e+05,1e+06,1e+07,
                           1e+08,1e+09,1e+02,1e+02,1e+03,1e+06,1e+06)
levels(data$PROPDMGEXP)
```

```
## [1] "1"      "0"      "10"     "100"    "1000"   "10000"  "1e+05"  "1e+06"
## [9] "1e+07"  "1e+08"  "1e+09"
```

```
data$PROPERTY_DAMAGE<-(data$PROPDMG*
                        as.numeric(as.character(data$PROPDMGEXP)))/1e+09
levels(data$CROPDGMGEXP)
```

```
## [1] ""  "?"  "0"  "2"  "B"  "k"  "K"  "m"  "M"
```

```
levels(data$CROPDGMGEXP)<-c(1,0,1,100,1e+09,1e+03,1e+03,1e+06,1e+06)
levels(data$CROPDGMGEXP)
```

```
## [1] "1"      "0"      "100"    "1e+09"  "1000"   "1e+06"
```

```
data$CROP_DAMAGE<-(data$CROPDMG*as.numeric(as.character(data$CROPDMGEXP)))/1e+09
```

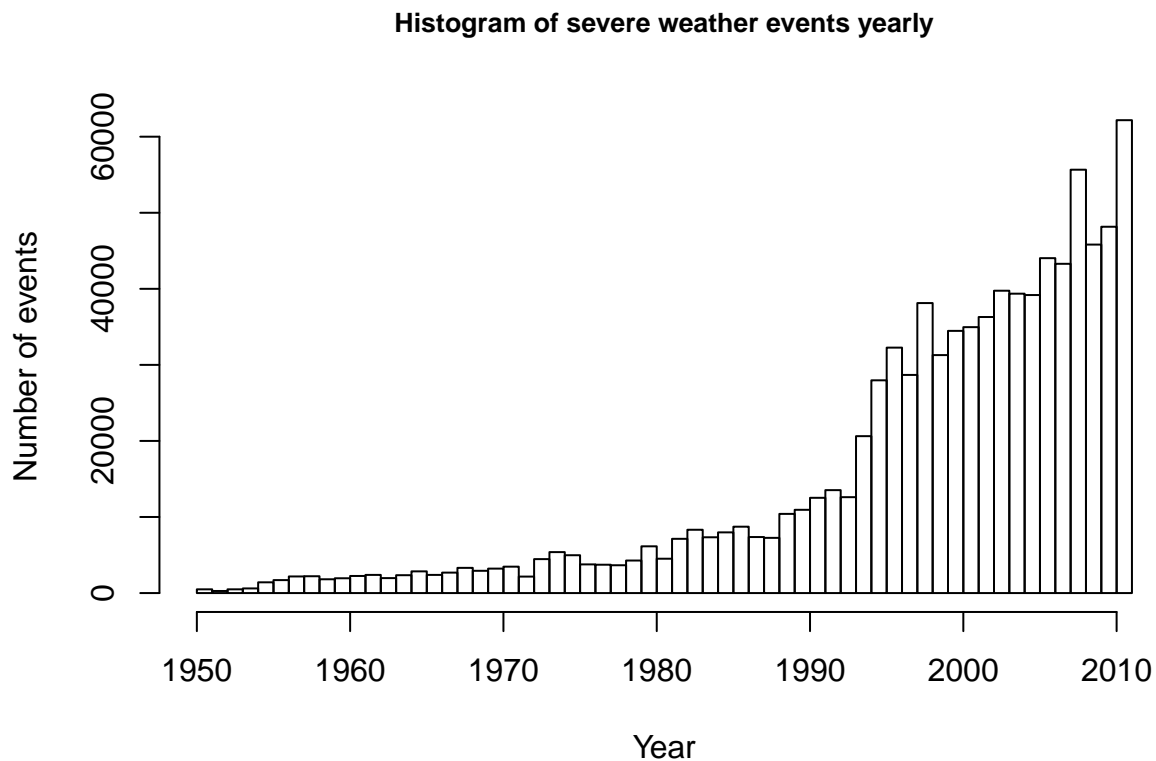
The events in the database start in the year 1950 and end in November 2011. In the earlier years of the database there are generally fewer events recorded, most likely due to a lack of good records. More recent years should be considered more complete.

Extract year from date and add a column “YEAR” in the data.

```
data$YEAR<-as.Date(as.character(data$BGN_DATE),format="%m/%d/%Y")
data$YEAR<-as.numeric(format(data$YEAR,"%Y"))
```

Make a histogram of years when physical disaster occurred.

```
hist(data$YEAR,50,xlab="Year",main="Histogram of severe weather events yearly",
     cex.main=0.85,ylab="Number of events")
```



Based on the above histogram, we see that the number of events tracked starts to significantly increase around 1993-1994. So we use the subset of the data from 1993 to 2011 to get most out of good records.

Subset the data from 1993 to 2011.

```
tidyData<-subset(data,data$YEAR>=1993)
```

The names of the columns/variables

```
names(tidyData)
```

```
## [1] "STATE_" "BGN_DATE" "BGN_TIME"
## [4] "TIME_ZONE" "COUNTY" "COUNTYNAME"
## [7] "STATE" "EVTYPE" "BGN_RANGE"
## [10] "BGN_AZI" "BGN_LOCATI" "END_DATE"
## [13] "END_TIME" "COUNTY_END" "COUNTYENDN"
## [16] "END_RANGE" "END_AZI" "END_LOCATI"
## [19] "LENGTH" "WIDTH" "F"
## [22] "MAG" "FATALITIES" "INJURIES"
## [25] "PROPDMG" "PROPDMGEXP" "CROPDMG"
## [28] "CROPDMGEXP" "WFO" "STATEOFFIC"
## [31] "ZONENAMES" "LATITUDE" "LONGITUDE"
## [34] "LATITUDE_E" "LONGITUDE_" "REMARKS"
## [37] "REFNUM" "PROPERTY_DAMAGE" "CROP_DAMAGE"
## [40] "YEAR"
```

## Results

### Impact on Public Health

```
fatalities<-aggregate(FATALITIES~EVTYPE,data=tidyData,FUN=sum)
fatalities<-fatalities[order(fatalities$FATALITIES,decreasing=TRUE),]
fatalities<-fatalities[1:10,]
row.names(fatalities)<-seq(1:10)
fatalities
```

```
##          EVTYPE FATALITIES
## 1 EXCESSIVE HEAT      1903
## 2     TORNADO      1621
## 3   FLASH FLOOD      978
## 4         HEAT      937
## 5   LIGHTNING      816
## 6       FLOOD      470
## 7   RIP CURRENT      368
## 8   HIGH WIND      248
## 9   TSTM WIND      241
## 10  AVALANCHE      224
```

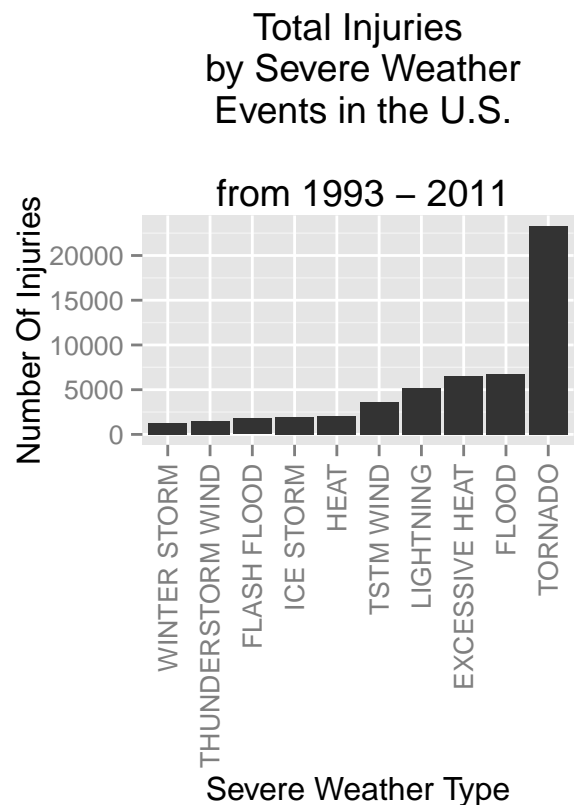
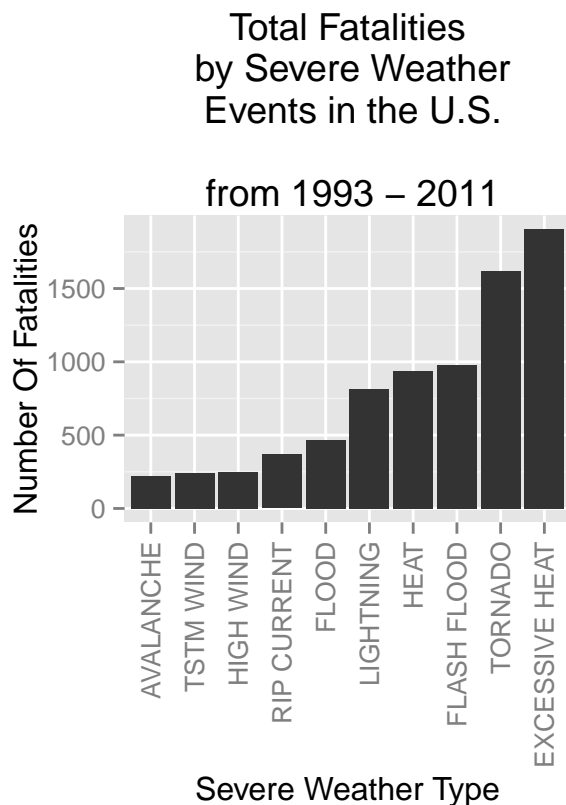
```
injuries<-aggregate(INJURIES~EVTYPE,data=tidyData,FUN=sum)
injuries<-injuries[order(injuries$INJURIES,decreasing=TRUE),]
injuries<-injuries[1:10,]
row.names(injuries)<-seq(1:10)
injuries
```

```
##          EVTYPE INJURIES
## 1     TORNADO    23310
## 2       FLOOD    6789
## 3 EXCESSIVE HEAT    6525
## 4   LIGHTNING    5230
## 5   TSTM WIND     3631
```

```
## 6          HEAT      2100
## 7      ICE STORM    1975
## 8      FLASH FLOOD  1777
## 9 THUNDERSTORM WIND  1488
## 10     WINTER STORM  1321
```

```
plotFatalities<-ggplot(fatalities,aes(reorder(EVTYPE,FATALITIES),FATALITIES))+
  geom_bar(stat="identity")+
  theme(axis.text.x=element_text(angle=90,hjust=1,vjust=0.5))+
  labs(x="Severe Weather Type",y="Number Of Fatalities",
       title="Total Fatalities\n by Severe Weather\n Events in the U.S.\n \n from 1993 - 2011")
plotInjuries<-ggplot(injuries,aes(reorder(EVTYPE,INJURIES),INJURIES))+
  geom_bar(stat="identity")+
  theme(axis.text.x=element_text(angle=90,hjust=1,vjust=0.5))+
  labs(x="Severe Weather Type",y="Number Of Injuries",
       title="Total Injuries\n by Severe Weather\n Events in the U.S.\n \n from 1993 - 2011")

grid.arrange(plotFatalities, plotInjuries, ncol = 2)
```



Based on the above Barplots, we find that excessive heat and tornado cause most fatalities. Tornado causes most injuries in the United States from 1993 to 2011.

### Impact on Economy

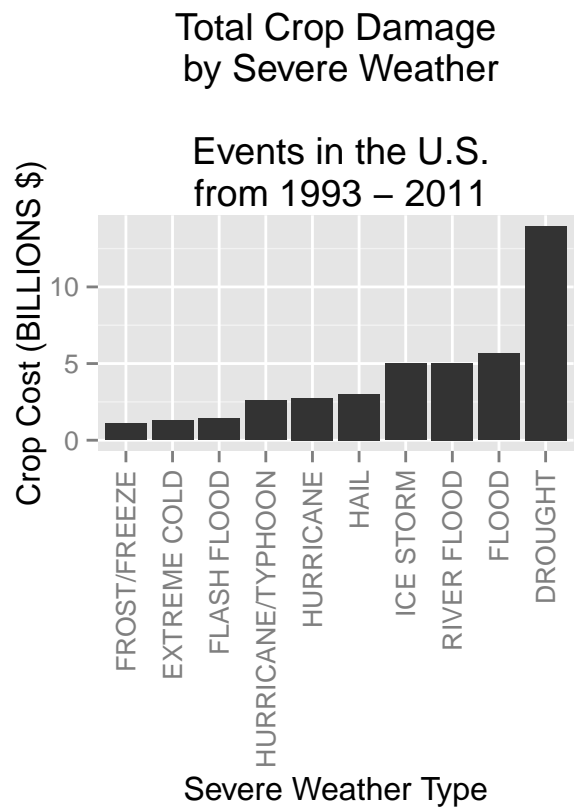
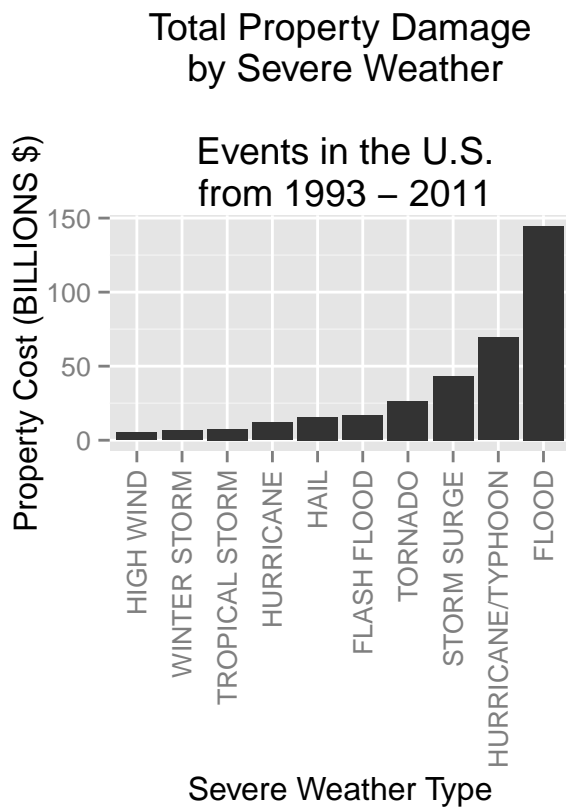
```
property_DMG<-aggregate(PROPERTY_DAMAGE~EVTYPE,data=tidyData,FUN=sum)
property_DMG<-property_DMG[order(property_DMG$PROPERTY_DAMAGE,decreasing=TRUE),]
property_DMG<-property_DMG[1:10,]
row.names(property_DMG)<-seq(1:10)
names(property_DMG)[2]<-"PROPERTY_DAMAGE(billions $)"
property_DMG
```

```
##           EVTYPE PROPERTY_DAMAGE(billions $)
## 1          FLOOD           144.657710
## 2 HURRICANE/TYPHOON           69.305840
## 3          STORM SURGE           43.323536
## 4          TORNADO           26.349182
## 5          FLASH FLOOD           16.822674
## 6           HAIL           15.735268
## 7          HURRICANE           11.868319
## 8    TROPICAL STORM            7.703891
## 9          WINTER STORM           6.688497
## 10         HIGH WIND            5.270046
```

```
crop_DMG<-aggregate(CROP_DAMAGE~EVTYPE,data=tidyData,FUN=sum)
crop_DMG<-crop_DMG[order(crop_DMG$CROP_DAMAGE,decreasing=TRUE),]
crop_DMG<-crop_DMG[1:10,]
row.names(crop_DMG)<-seq(1:10)
names(crop_DMG)[2]<-"CROP_DAMAGE(billions $)"
crop_DMG
```

```
##           EVTYPE CROP_DAMAGE(billions $)
## 1          DROUGHT           13.972566
## 2          FLOOD            5.661968
## 3    RIVER FLOOD            5.029459
## 4          ICE STORM            5.022113
## 5           HAIL            3.025954
## 6          HURRICANE            2.741910
## 7 HURRICANE/TYPHOON            2.607873
## 8          FLASH FLOOD            1.421317
## 9          EXTREME COLD            1.292973
## 10         FROST/FREEZE            1.094086
```

```
plotProperty<-ggplot(property_DMG,aes(reorder(EVTYPE,property_DMG[,2]),
                                         property_DMG[,2]))+
  geom_bar(stat="identity")+
  theme(axis.text.x=element_text(angle=90,hjust=1,vjust=0.5))+
  labs(x="Severe Weather Type",y="Property Cost (BILLIONS $)",
       title="Total Property Damage\n by Severe Weather\n Events in the U.S.\n from 1993 - 2011")
plotCrop<-ggplot(crop_DMG,aes(reorder(EVTYPE,crop_DMG[,2]),crop_DMG[,2]))+
  geom_bar(stat="identity")+
  theme(axis.text.x=element_text(angle=90,hjust=1,vjust=0.5))+
  labs(x="Severe Weather Type",y="Crop Cost (BILLIONS $)",
       title="Total Crop Damage\n by Severe Weather\n Events in the U.S.\n from 1993 - 2011")
grid.arrange(plotProperty, plotCrop, ncol = 2)
```



Based on the above barplots, across the United States, flood and typhoon have caused the greatest damage to properties. Drought and flood have caused the greatest damage to crops.